# Simple Random Walk on a Graph

Math 365 Tanya Leise

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Submit your completed lab to Moodle as a pdf or html document.

The objective of these exercises is to study a random walk on a small graph through a combination of simulations and analysis using R.

**Problem set-up** Consider the simple random walk on the graph shown below. We define a Markov chain which moves from vertex to vertex by randomly following an edge to an adjacent vertex. For example,

$$\mathbb{P}\{X_{n+1} = A \mid X_n = D\} = 1/2$$

$$\mathbb{P}\{X_{n+1} = C \mid X_n = A\} = 1/3$$

$$\mathbb{P}\{X_{n+1} = B \mid X_n = D\} = 0$$

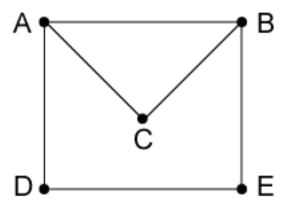


Figure 1: Graph

Determine the transition matrix **P** for this Markov chain with states  $\{A, B, C, D, E\}$ .

## library(expm)

## Loading required package: Matrix

##

## Attaching package: 'expm'

```
## The following object is masked from 'package:Matrix':
##
##
       expm
P \leftarrow matrix(0,5,5)
P[1,] \leftarrow c(0,1/3,1/3,1/3,0)
P[2,] \leftarrow c(1/3,0,1/3,0,1/3)
P[3,] \leftarrow c(1/2,1/2,0,0,0)
P[4,] \leftarrow c(1/2,0,0,0,1/2)
P[5,] \leftarrow c(0,1/2,0,1/2,0)
rownames(P)<-c('A','B','C','D','E')
colnames(P)<-c('A','B','C','D','E')</pre>
              Α
                         В
                                    C
                                               D
## A 0.0000000 0.3333333 0.3333333 0.0000000
## B 0.3333333 0.0000000 0.3333333 0.0000000 0.3333333
## C 0.5000000 0.5000000 0.0000000 0.0000000 0.0000000
## D 0.5000000 0.0000000 0.0000000 0.0000000 0.5000000
## E 0.0000000 0.5000000 0.0000000 0.5000000 0.0000000
```

## Analysis of the Markov chain

1. Calculate the invariant probability vector  $\bar{\pi}$ . In the long run, about what fraction of the time is spent at vertex A?

Sol: 0.25

## [1] 0.25

```
P %^% 100
##
                                           Ε
        Α
## A 0.25 0.25 0.1666667 0.1666667 0.1666667
## B 0.25 0.25 0.1666667 0.1666667 0.1666667
## C 0.25 0.25 0.1666667 0.1666667 0.1666667
## D 0.25 0.25 0.1666667 0.1666667 0.1666667
## E 0.25 0.25 0.1666667 0.1666667 0.1666667
r <- eigen(t(P))
V<-r$vectors
pibar <- V[,1]/sum(V[,1])
pibar
## [1] 0.2500000 0.2500000 0.1666667 0.1666667
ans1 <- pibar[1]</pre>
ans1
```

2. Suppose the random walk begins at vertex A. What is the expected number of steps until the walk returns to A?

```
ans2 <- 1/pibar[1]
ans2</pre>
```

```
## [1] 4
```

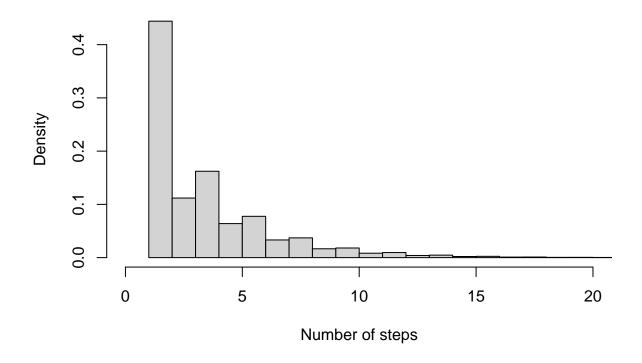
We can use simulations to generate the distribution of return times, as the expected value doesn't tell the full story. First define a function that can randomly step from vertex to vertex:

Next define a function that finds the return time to a desired vertex for each simulation:

```
waitingtime <- function(vertex) {
  x <- vertex
  for (j in 1:10000) {
    x <- takestep(x)
    ifelse(x==vertex, return(j),-1) # takes steps until hits vertex again
    } }</pre>
```

Run many simulations and plot a histogram of the return times:

## Mean first return time is 3.99415



3. Suppose the random walk begins at vertex C. What is the expected number of steps until the walker reaches A? Hint: Make A an absorbing state, then calculate matrices  $\mathbf{Q}$  and  $\mathbf{M}$ .

```
Q<-P[2:5,2:5]
##
       В
                 С
                     D
## B 0.0 0.3333333 0.0 0.3333333
## C 0.5 0.0000000 0.0 0.0000000
## D 0.0 0.0000000 0.0 0.5000000
## E 0.5 0.0000000 0.5 0.0000000
M<-solve(diag(4)-Q)
                       С
##
             В
                                  D
                                            Ε
## B 1.6363636 0.5454545 0.3636364 0.7272727
## C 0.8181818 1.2727273 0.1818182 0.3636364
## D 0.5454545 0.1818182 1.4545455 0.9090909
## E 1.0909091 0.3636364 0.9090909 1.8181818
ans3 < -sum(M[2,])
ans3
```

## ## [1] 2.636364

4. Suppose the random walk begins at vertex C. What is the expected number of visits to B before the walker reaches A?

```
M[2,1]
```

## ## [1] 0.8181818

5. Suppose the random walk begins at vertex B. What is the probability that the walker reaches A before C?

```
P[1,] < -c(1,0,0,0,0)
P[3,] < -c(0,0,1,0,0)
##
                           С
                                D
                                          Ε
             Α
## A 1.0000000 0.0 0.0000000 0.0 0.0000000
## B 0.3333333 0.0 0.3333333 0.0 0.3333333
## C 0.0000000 0.0 1.0000000 0.0 0.0000000
## D 0.5000000 0.0 0.0000000 0.0 0.5000000
## E 0.0000000 0.5 0.0000000 0.5 0.0000000
P_ac_absorbing_asymptotic <- P % ^% 1000
P_ac_absorbing_asymptotic
                                                                   Ε
##
                           В
                                      C
                                                    D
             Α
## A 1.0000000 0.000000e+00 0.0000000 0.000000e+00
                                                       0.000000e+00
## B 0.5714286 3.136456e-191 0.4285714 3.136456e-191
                                                       0.000000e+00
```

```
P_ac_absorbing_asymptotic[2,1]
```

0.000000e+00

## C 0.0000000 0.000000e+00 1.0000000 0.000000e+00

## D 0.8571429 4.704683e-191 0.1428571 4.704683e-191 0.000000e+00 ## E 0.7142857 0.000000e+00 0.2857143 0.000000e+00 7.841139e-191

## [1] 0.5714286